

DESIGN SENSITIVITY ANALYSIS: APPLICATIONS

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- Overview
- Tangent matrix
- Multi-level Newton method: Coupled analyses
- Multi-level Newton method: Domain decomposition
- Optimization environment

Introduction

Optimization problem

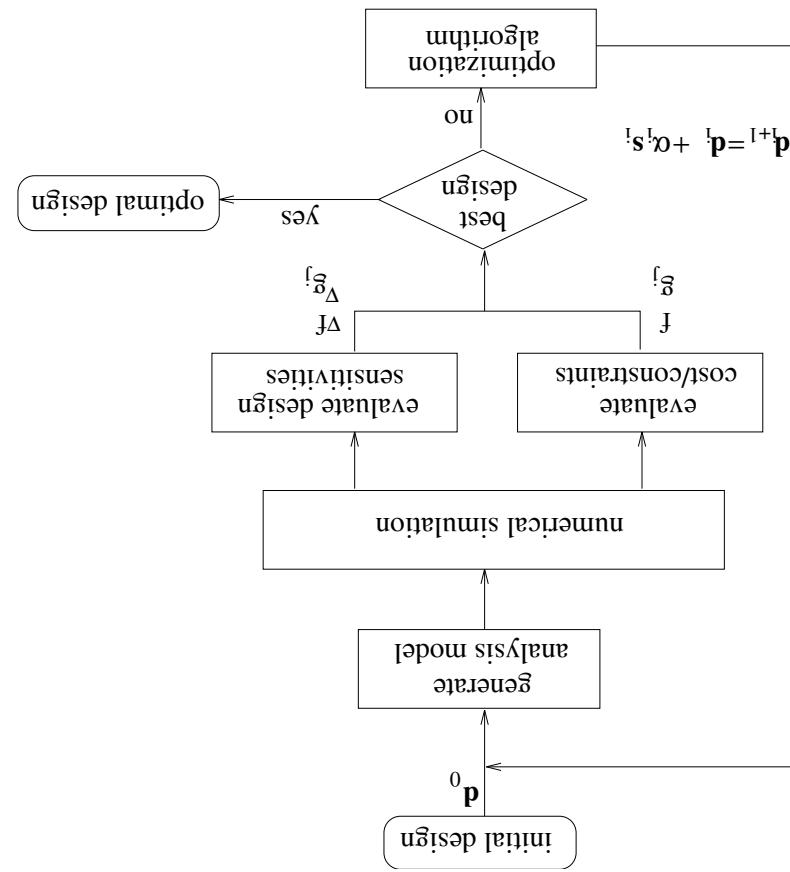
$$\min_{\mathbf{p}} F(\mathbf{p}) \text{ such that } G_i(\mathbf{p}) \geq 0, i = 1, 2, \dots, n_g$$

- F cost function

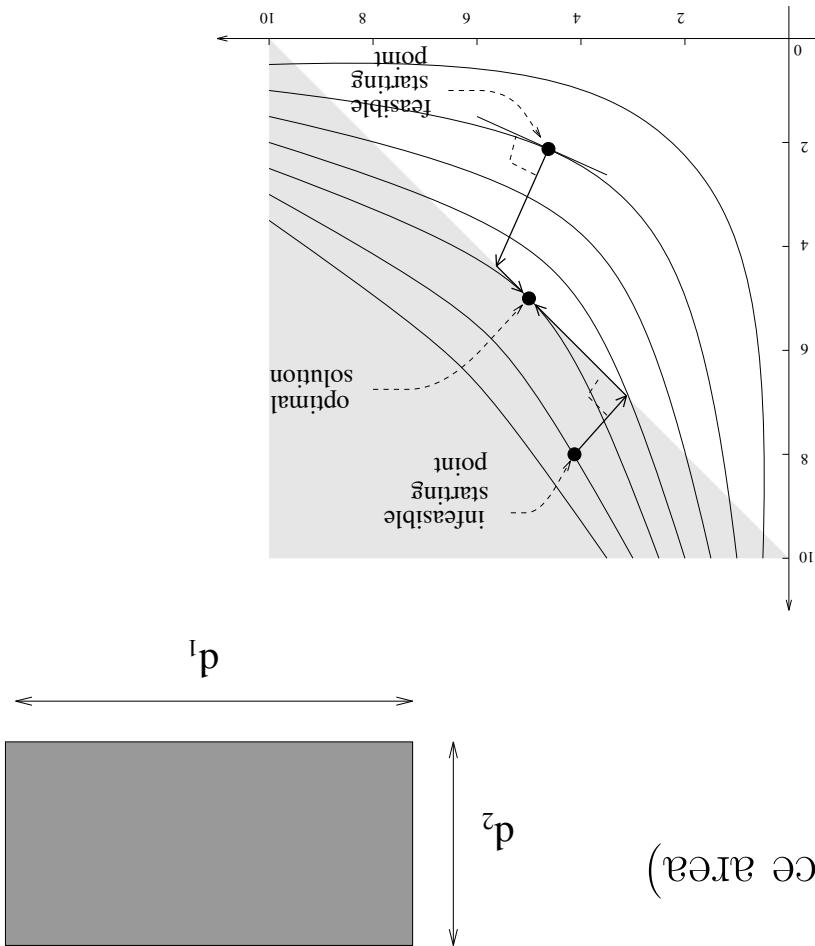
- G_i inequality constraint

Nonlinear Programming Caveats

- Continuous design space
- Deterministic analyses
- Smooth cost and constraint functions
- Local minima



Optimization algorithm



- Maximize objective function (surface area)
- Constraints (on perimeter)

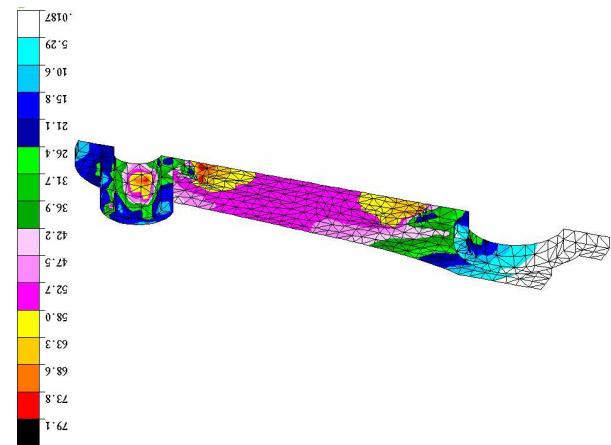
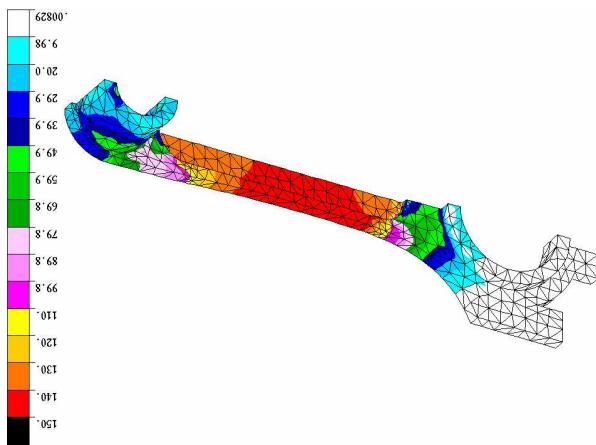
$$F = d_1 d_2$$

$$G = 2(d_1 + d_2) - 20 \leq 0$$

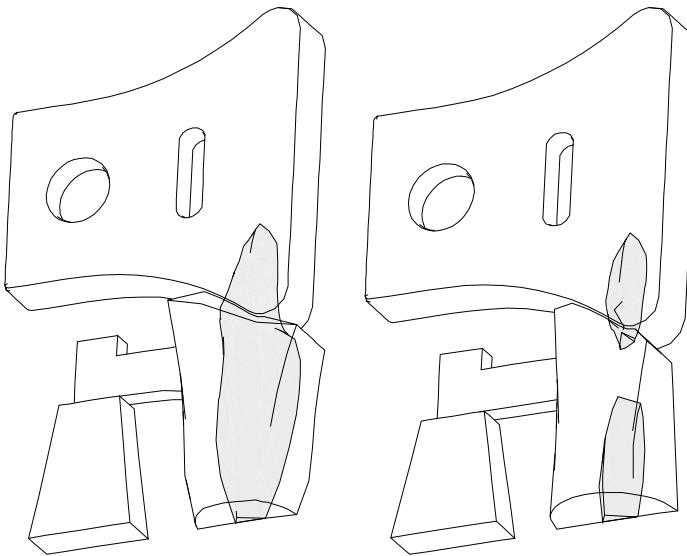
- Side Constraints

$$d_i < 0$$

Example: sheet design

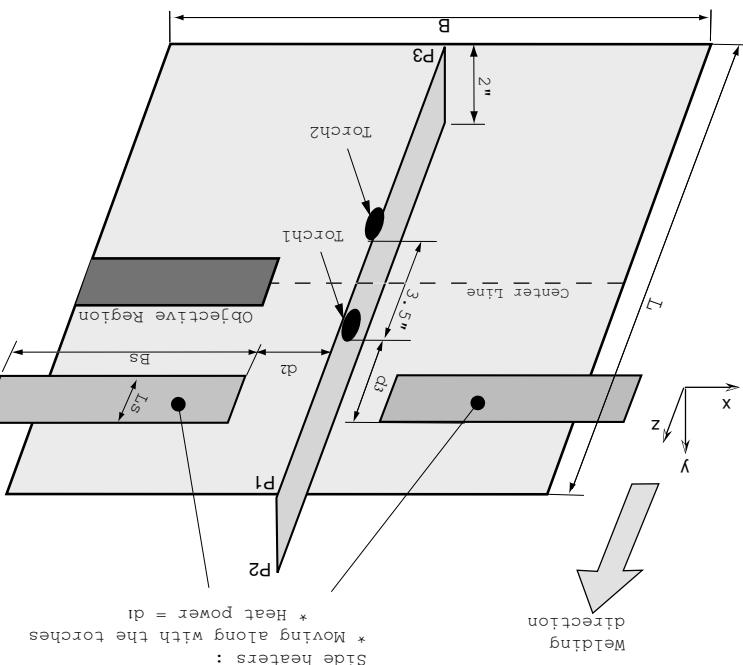
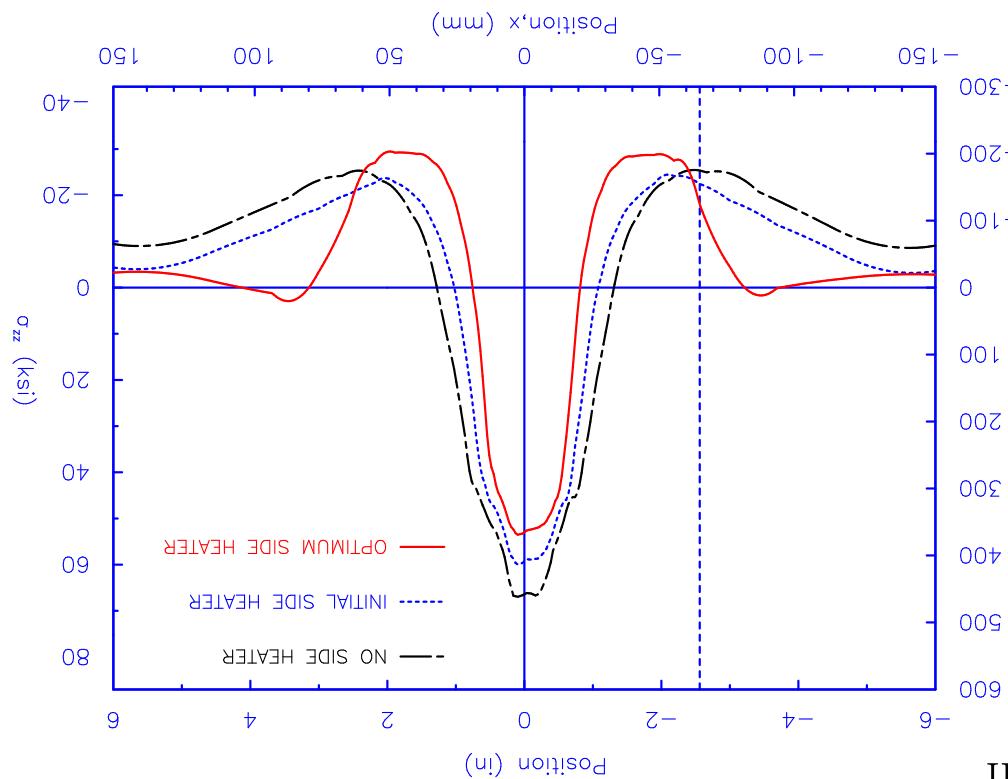


Connecting rod shape optimization



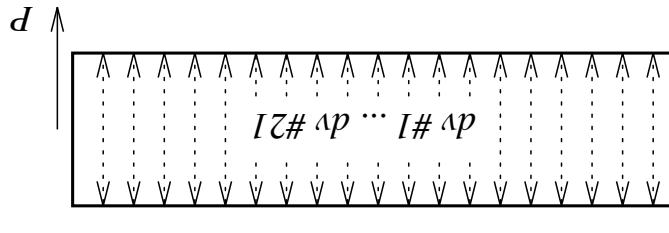
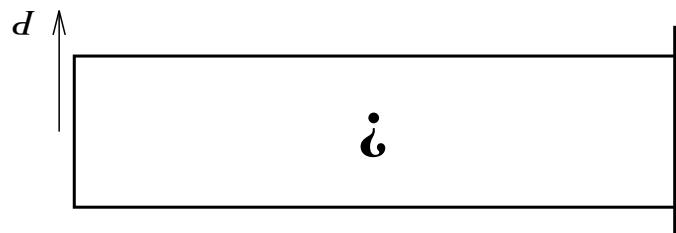
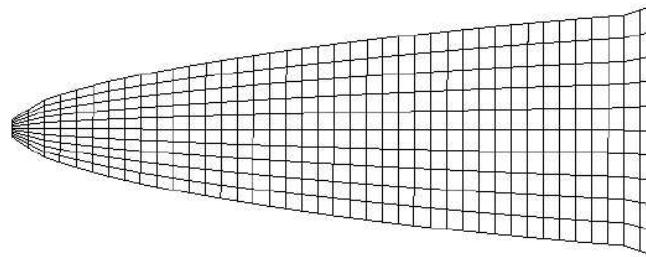
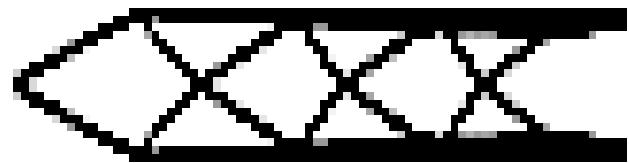
- Thermal conduction
- Enforce directional solidification
- Minimize riser volume

Casting optimization



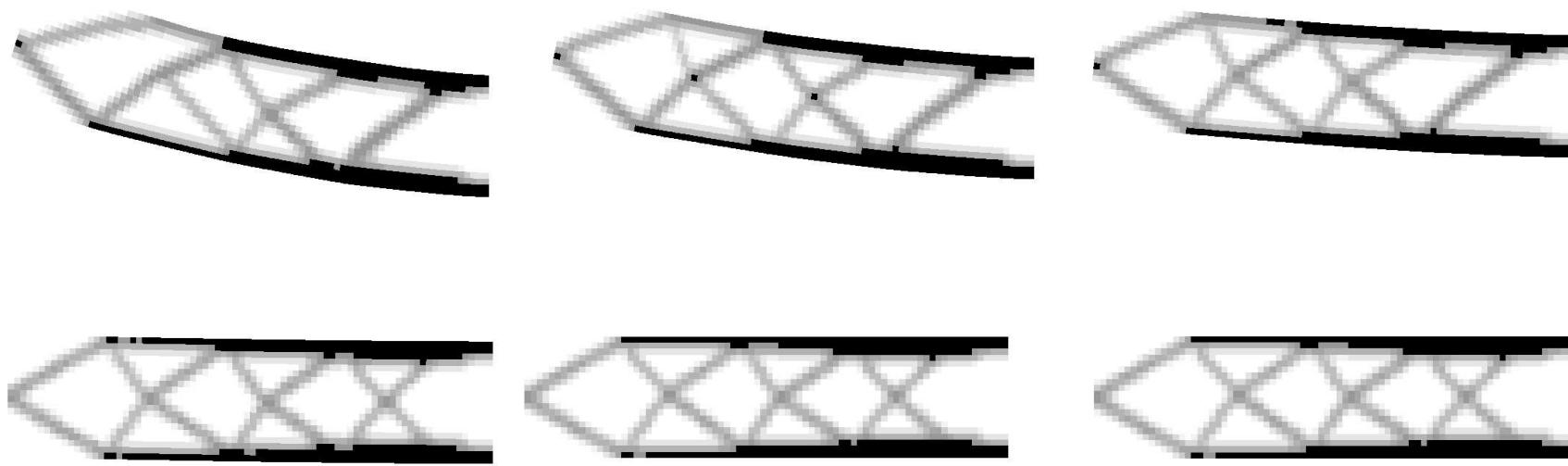
- Design thermal tensiioning process
- Minimize longitudinal residual stress away from heat effected zone
- Optimize heater power and location

VWeld optimization

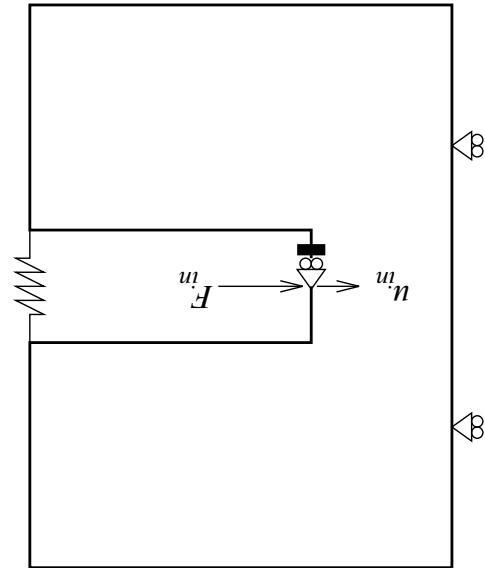
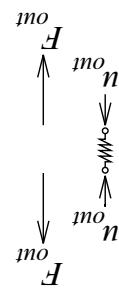
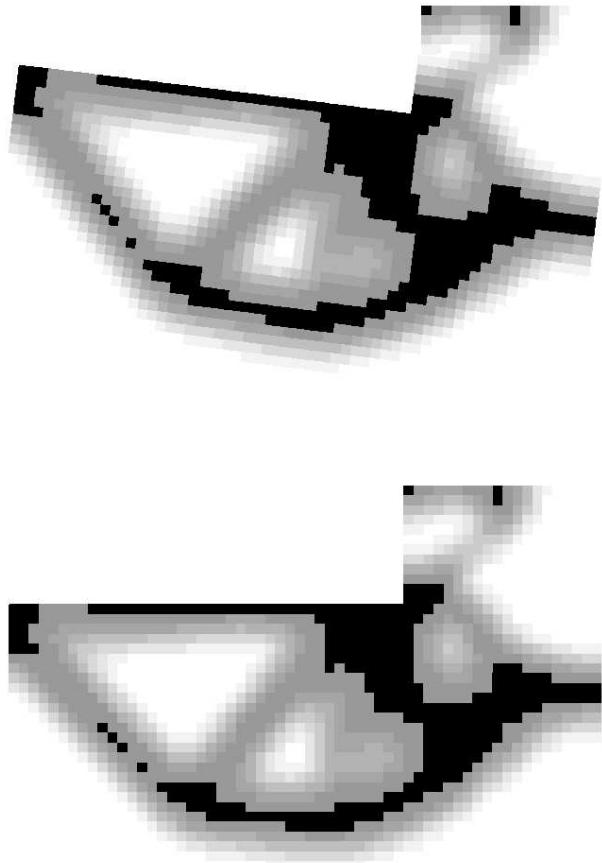


- Many design parameters
- Distribute material

Topology optimization

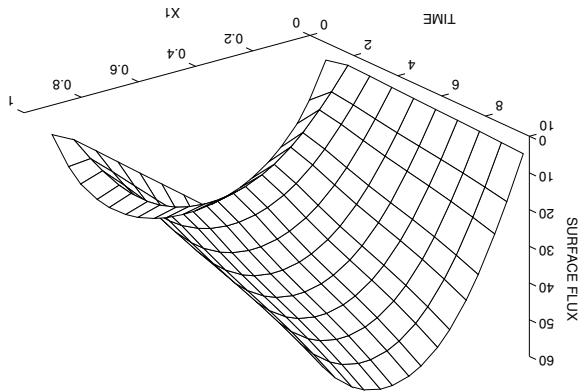


Topology optimization: Load dependent designs

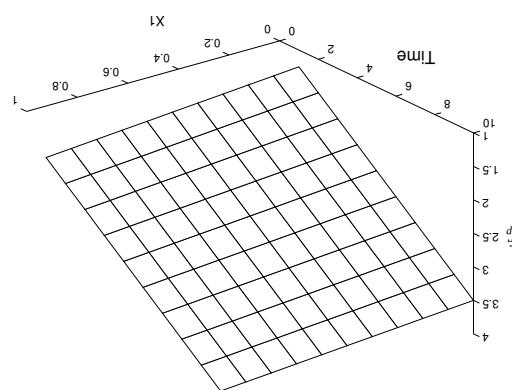


Topology optimization: Complicated mechanisms

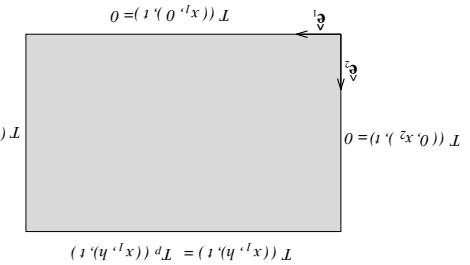
Experimental Flux



Target distribution



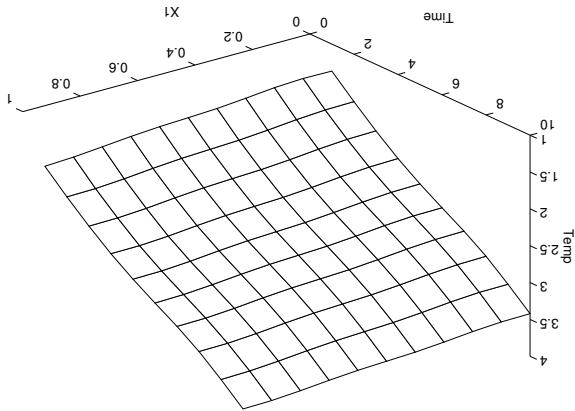
Plate



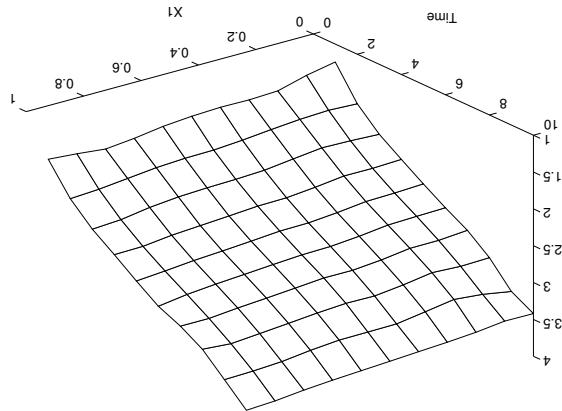
- Ill-posed \Leftarrow Need regularization
- 81 variables
- Minimize Error, $E = \int \frac{1}{2} (b_{data} - b)^2 dx$
- Find T^d

Identification problem: example

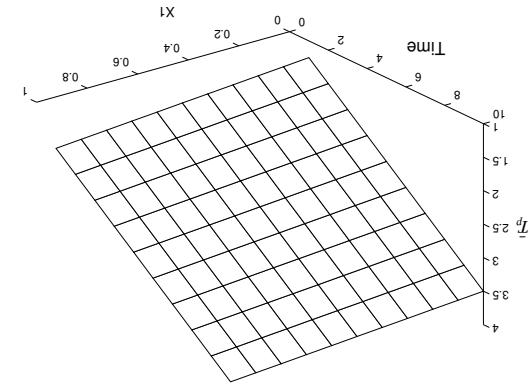
1 iteration
w/ regularization



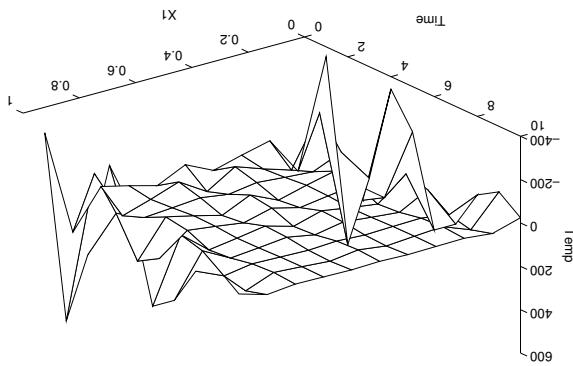
≈ 40 iterations
w/ regularization



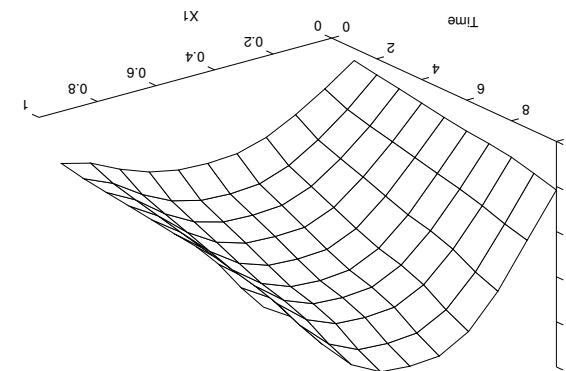
Target distribution



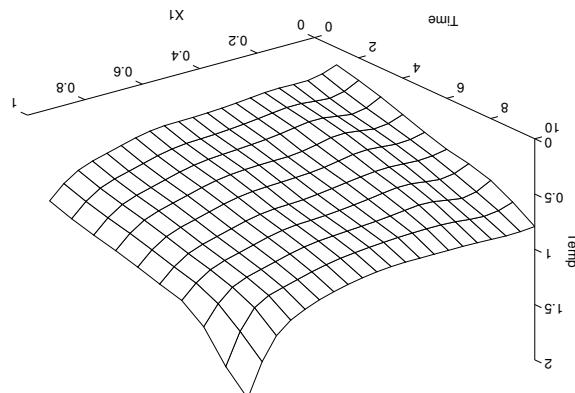
Newton's



BFGS



Initial distribution



- $$\underbrace{\left\{ \mathbf{X} \frac{\mathbf{n}_d}{\partial \mathbf{R}} - \frac{\mathbf{n}_d}{\partial \mathbf{G}^T} \right\} \frac{\partial d}{D^T} + \frac{\partial d}{\partial \mathbf{G}} \mathbf{X} - \frac{\partial d}{\partial \mathbf{G}} = 0}_{\text{--- Adjoint: } \frac{\partial F}{\partial d} = \frac{\partial G}{\partial d}}$$
- $$\frac{\partial \mathbf{R}}{\partial \mathbf{u}} \frac{\partial \mathbf{u}}{\partial d} + \frac{\partial \mathbf{R}}{\partial d} = \mathbf{0}$$
- $$\frac{\partial \nabla}{\partial d} \approx \frac{G(\mathbf{u}(p) + p\nabla + p^2(p) - G(\mathbf{u}(p))}{(p^2(p)\mathbf{u}(p) + p^2(p)d - G(\mathbf{u}(p)))}$$
- Sensitivity $\frac{\partial F}{\partial d} = \frac{\partial G}{\partial d} + \frac{\partial G}{\partial \mathbf{u}} \frac{\partial \mathbf{u}}{\partial d}$
 - Generalized response functional $F(d) = G(\mathbf{u}(d), d)$
- Newton update, $\left[\frac{\partial \mathbf{R}}{\partial \mathbf{u}}(\mathbf{u}_I) \right] \nabla \mathbf{u} = -\mathbf{R}(\mathbf{u}_I)$
 - Evaluate residual, $\mathbf{R}(\mathbf{u}_I) \stackrel{?}{=} \mathbf{0}$
- Analysis

Sensitivity analysis

- Use automatic differentiation, symbolic algebra or finite difference: $\frac{\partial \mathbf{R}}{\partial \mathbf{d}}$
- Coupled analyses: OK
- Transient analyses: Implicit time integration preferred
- Eliminate truncation $\not\approx$ round-off errors
- n linear pseudo or m linear adjoint problems
- Direct solvers preferred

Sensitivity analysis: Notes

$$\frac{p\varrho}{\partial G} + \frac{d}{\mathbf{n}} \frac{\mathbf{n}_u \varrho}{D_u \partial G} = \frac{d}{D}$$

$$(p, (p) \mathbf{n}_u) \mathbf{G} = F(p)$$

- Sensitivity Analysis:

$$(\mathbf{n}_{I-u}, \mathbf{n}_u) \mathbf{R}_u - = \mathbf{n} \nabla (\mathbf{n}_{I-u}, \mathbf{n}_u) \frac{\mathbf{n}_u \varrho}{\mathbf{R}_u \varrho}$$

- Newton-Raphson

$$\mathbf{0} = (\mathbf{n}_{I-u}, \mathbf{n}_u) \mathbf{R}_u$$

- Residual at time t_n

Incremental Form:

Nonlinear transient sensitivity analysis

• Direct differentiation:

Direct differentiation:

- Residual at time t_u

- Differentiate

$$\mathbf{0} = \frac{p\varrho}{\varrho u \mathbf{B}} + \frac{d}{\mathbf{B} \cdot \mathbf{n}} \frac{\mathbf{n}_{L-u} \varrho}{D} + \frac{d}{\mathbf{B} \cdot \mathbf{n}} \frac{\mathbf{n}_u \varrho}{D}$$

- Pseudo problem: $\frac{D^d}{D_u \mathbf{n}}$

- Compute $\frac{D^d}{D_u \mathbf{n}}$ simultaneously with \mathbf{n}_u .

$$\left(\frac{p\varrho}{\varrho u \mathbf{B}} + \frac{d}{\mathbf{B} \cdot \mathbf{n}} \frac{\mathbf{n}_{L-u} \varrho}{D} \right) - = \frac{d}{\mathbf{B} \cdot \mathbf{n}} \frac{\mathbf{n}_u \varrho}{D}$$

$$\frac{pD}{\mathbf{n}_0} \frac{\partial \varrho}{\partial \mathbf{B}} - \frac{p\varrho}{\mathbf{B}^T \frac{\partial \mathbf{B}}{\partial \mathbf{B}}} \sum_u^{l=j} - \frac{p\varrho}{\mathcal{G}} \equiv \frac{Dd}{DF}$$

- Explicit component

$$\left(\frac{p\varrho}{\mathbf{n}_{l-j}} + \frac{pD}{\mathbf{n}_{l-j} D_j} \frac{\partial \varrho}{\partial \mathbf{B}} + \frac{pD}{\mathbf{n}_j} \frac{\partial \varrho}{\partial \mathbf{B}} \right) \mathbf{B}^T \sum_u^{l=j} - \frac{p\varrho}{\mathcal{G}} + \frac{Dd}{\mathbf{n}_j \mathcal{G}} = \frac{Dd}{DF}$$

- Augment functional

Adjoint method:

$$\begin{aligned}
& \left(\frac{pD}{n_u D} \frac{\mathbf{n}_u \theta}{D u G} \frac{pD}{n_u D} \frac{\mathbf{n}_u \theta}{D u R} \mathbf{L} \boldsymbol{\chi}_u \right) - \\
& \left(\frac{pD}{n_{I-u} D} \frac{\mathbf{n}_{I-u} \theta}{D u R} \mathbf{L} \boldsymbol{\chi}_u + \frac{pD}{n_{I-u} D} \frac{\mathbf{n}_{I-u} \theta}{D u R} \mathbf{L} \boldsymbol{\chi}_{I-u} \right) - \\
& \dots \\
& \left(\frac{pD}{n_2 D} \frac{\mathbf{n}_2 \theta}{D u R} \mathbf{L} \boldsymbol{\chi}_3 + \frac{pD}{n_2 D} \frac{\mathbf{n}_2 \theta}{D u R} \mathbf{L} \boldsymbol{\chi}_2 \right) - \\
& \left(\frac{pD}{n_1 D} \frac{\mathbf{n}_1 \theta}{D u R} \mathbf{L} \boldsymbol{\chi}_1 + \frac{pD}{n_1 D} \frac{\mathbf{n}_1 \theta}{D u R} \mathbf{L} \boldsymbol{\chi}_I \right) - = \frac{D_F}{D_I}
\end{aligned}$$

- Implicit component

Adjoint method (cont)

- Requires much storage or recompilation

$$\mathbf{Y}_2 \left(\frac{\mathbf{n}_L \ell}{\mathbf{e}_2 \mathbf{R}} \right) - = \mathbf{Y}_1 \left(\frac{\mathbf{n}_L \ell}{\mathbf{e}_1 \mathbf{R}} \right) \quad : 1$$

$$\mathbf{Y}_3 \left(\frac{\mathbf{n}_2 \ell}{\mathbf{e}_3 \mathbf{R}} \right) - = \mathbf{Y}_2 \left(\frac{\mathbf{n}_2 \ell}{\mathbf{e}_2 \mathbf{R}} \right) \quad : 2$$

...

$$\mathbf{Y}_u \left(\frac{\mathbf{n}_{L-u} \ell}{\mathbf{e}_u \mathbf{R}} \right) - = \mathbf{Y}_{L-u} \left(\frac{\mathbf{n}_{L-u} \ell}{\mathbf{e}_{L-u} \mathbf{R}} \right) : L-u$$

$$\frac{\mathbf{n}_u \ell}{\mathcal{C}\ell} = \mathbf{Y}_u \left(\frac{\mathbf{n}_u \ell}{\mathbf{R}_u \ell} \right) : u$$

- Solve adjoint problems

Adjoint method (cont)

$$\frac{u \mathbf{n} \cdot \boldsymbol{\sigma}}{\mathbf{x} \cdot \boldsymbol{\sigma}} = \begin{Bmatrix} \frac{u \mathbf{n} \cdot \boldsymbol{\sigma}}{u \boldsymbol{\sigma} \cdot \boldsymbol{\sigma}} \\ \frac{u \mathbf{n} \cdot \boldsymbol{\sigma}}{u \boldsymbol{\sigma} \cdot \boldsymbol{\sigma}} \end{Bmatrix} \frac{(\boldsymbol{\alpha}_n, \boldsymbol{\sigma}_n) \cdot \boldsymbol{\sigma}}{\mathbf{x} \cdot \boldsymbol{\sigma}}$$

- Differentiate to obtain

$$\mathbf{r}((\boldsymbol{\sigma}_n, \boldsymbol{\alpha}_n), (\mathbf{u}_n, \boldsymbol{\sigma}_{n-1}, \boldsymbol{\alpha}_{n-1})) = \mathbf{0}$$
- Solve the residual equation
- Given \mathbf{u}_n , $\boldsymbol{\sigma}_{n-1}$ and $\boldsymbol{\alpha}_{n-1}$ compute $\boldsymbol{\sigma}_n$, $\boldsymbol{\alpha}_n$ and $\boldsymbol{\sigma}_n \cdot \boldsymbol{\sigma}_n / \mathbf{u}_n$
- E.g. stress and yield stress are implicit functions of displacement
- Secondary response is an implicit function of primary response
- Subiterations required to compute \mathbf{R}

Tangent matrix

- Finite difference

- Symbolic algebra, automatic differentiation

$$\frac{H\varrho}{\mathbf{x}\varrho} = \begin{Bmatrix} \frac{H\varrho}{L\varrho} \\ \frac{H\varrho}{s_\varrho\varrho} \end{Bmatrix} \left[\frac{(L^{s_\varrho})\varrho}{\mathbf{x}\varrho} \right] \quad \text{and} \quad \begin{bmatrix} C^d & T^- \\ \frac{L^{-f_L}}{\left(\frac{L^{-f_L}}{L^{-f_L}}\right)(1-k)} & 1 \end{bmatrix} = \frac{(L^{s_\varrho})\varrho}{\mathbf{x}\varrho}$$

$$\begin{Bmatrix} 1^- \\ 0 \end{Bmatrix} = \frac{H\varrho}{\mathbf{x}\varrho} \quad \begin{Bmatrix} H - T^{(s_\varrho - 1) + L^d} \\ \left(\frac{L^{-f_L}}{L^{-f_L}}\right) + 1 - s_\varrho \end{Bmatrix} = (H : (s_\varrho, L))\mathbf{x}$$

- Mushy zone relations: Scheil assumption
- Secondary response: solid volume fraction s and temperature T
- Primary response: enthalpy H

Phase change analysis

- Second evolutionary state variable χ : $\dot{\chi} = \frac{d\chi}{dt} = a'' \frac{L^b}{1 - \frac{L^b}{L}}$
- Strength evolution: $d\theta = \frac{JL}{\varepsilon} \left[\frac{L^b}{q} + \frac{s_\varepsilon}{\varepsilon} - 1 \right] \theta^o$
- Crystal flow stress: $\tau - \tau^a = \frac{u}{S(\varepsilon, T)} \frac{L^b}{L^b + L^a}$

Stage IV hardening w/ two state variable hybrid model

Parameter identification

- Identify material parameters and initial conditions

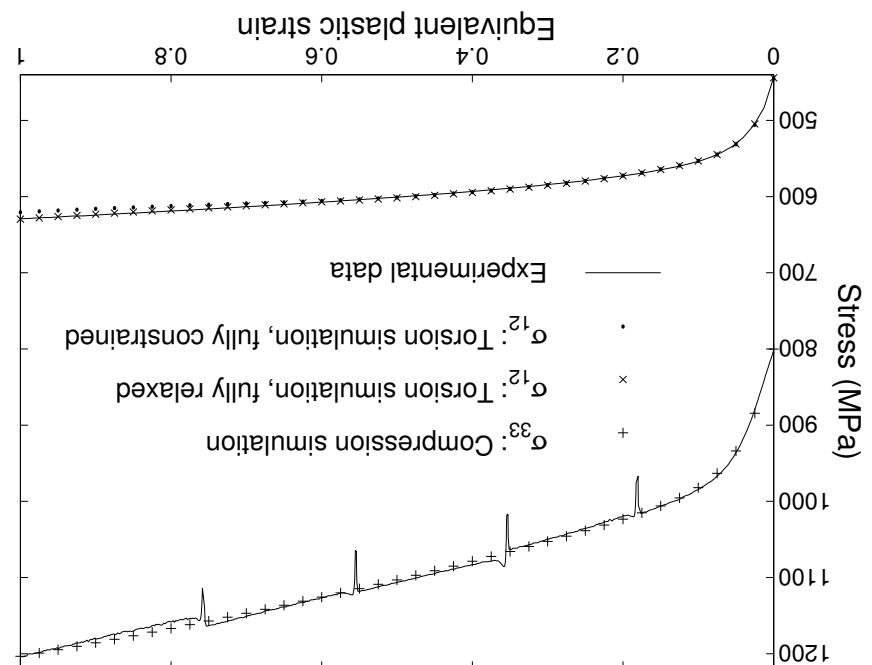
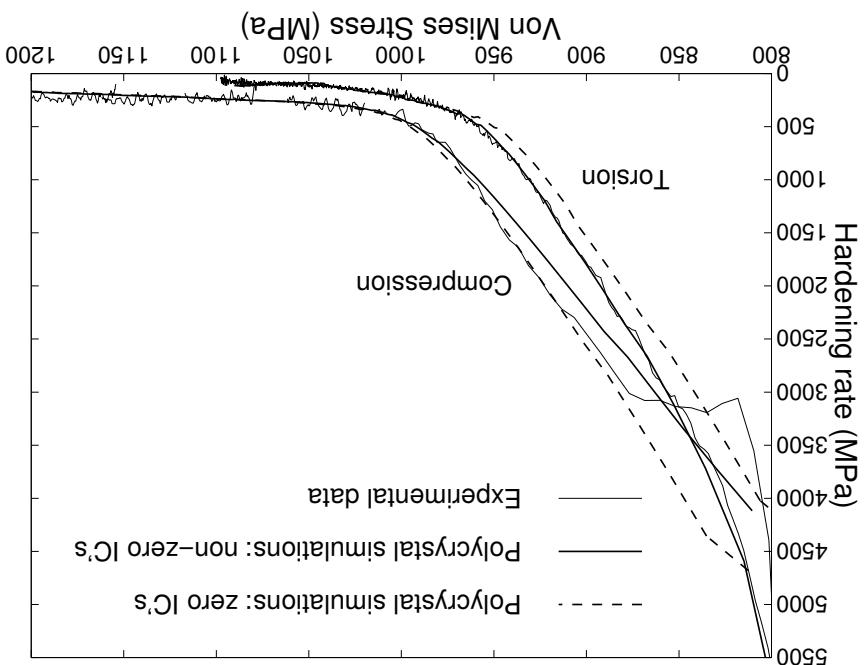
Material parameters

Initial conditions

$$[0 = J \left| \frac{\dot{x}}{x} \right. , 0 = J \left| \frac{\dot{z}}{z} \right. , \dots, 0 = J \left| \frac{\dot{L}}{L} \right.] = \dot{\boldsymbol{p}}$$

$$[o_3, o_2, o_1, \theta_o, c, u, \dot{L}^a, \dot{L}^{so}, \dot{\gamma}_{so}, \dot{\gamma}_{oe}, \dot{y}_{oe}] = \dot{\boldsymbol{p}}$$

$$\underbrace{\left(\frac{\frac{d}{dt} \tilde{J}_i}{\tilde{J}_i} - 1 \right) \sum_{j=1}^d}_{\text{Minimize}} = (\boldsymbol{p})_E \quad \boldsymbol{p} = [\boldsymbol{p}^d, \boldsymbol{p}^i]$$



HY100 Steel

- Linear analyses

- Matrix free parallelization

$$\begin{Bmatrix} {}^h s + [(x)h - ({}^x s \epsilon + x)h] \frac{\epsilon}{1} - \\ [h]x - ({}^h s \epsilon + h)x] \frac{\epsilon}{1} - {}^x s \end{Bmatrix} = \begin{Bmatrix} {}^h s + {}^x s (x)h - \\ {}^h s (h)x - {}^x s \end{Bmatrix} = \mathbf{s}(h, x) \mathbf{R}(x, y)$$

- Iteratively solve Newton update (GMRES, Ublk Solve)

$$\begin{Bmatrix} (x)h - h \\ [h]x - x \end{Bmatrix} - = \begin{Bmatrix} h \nabla \\ x \nabla \end{Bmatrix} \begin{bmatrix} 1 & -y(x) \\ (h)x - & 1 \end{bmatrix} = (h, x) \mathbf{R}(x, y) \mathbf{R}(x, y) = \nabla$$

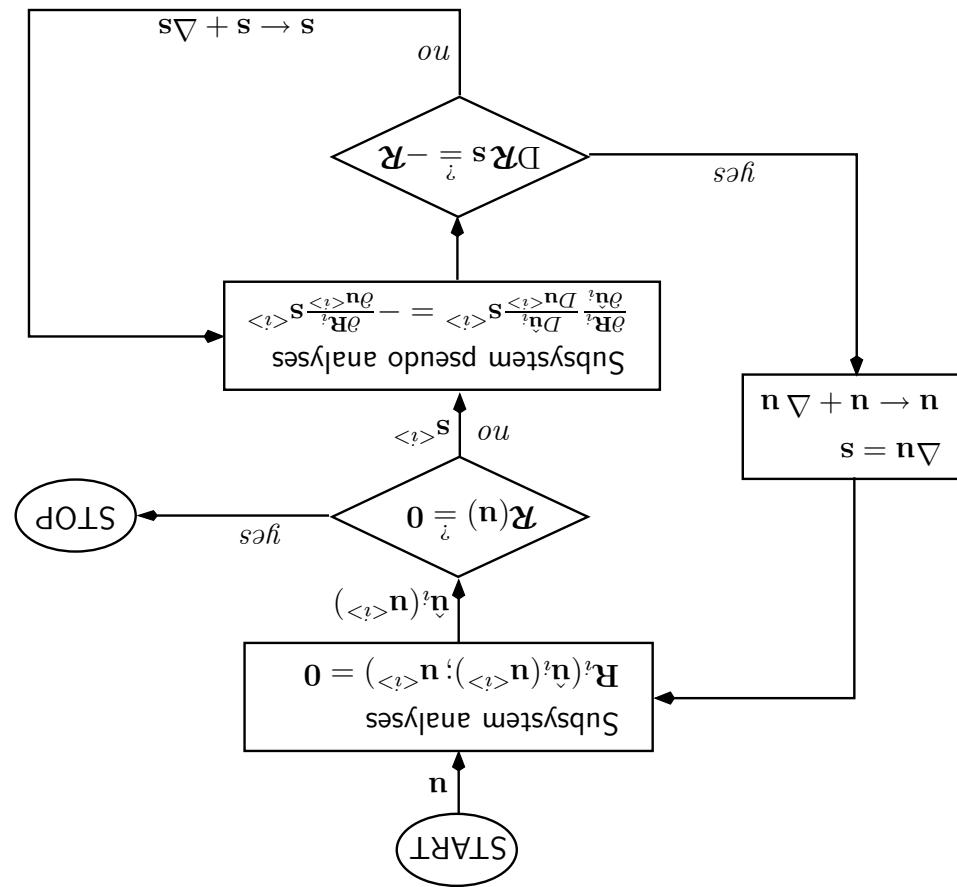
- Solve by Newton's method

$$\mathbf{0} = \begin{Bmatrix} (x)h - h \\ [h]x - x \end{Bmatrix} = (h, x) \mathbf{R}(x, y)$$

- Form a "self consistent" residual

- Coupled analyses: $f(x, y) = 0$ and $g(y) = 0$

Multi-level Newton method: Coupled analyses

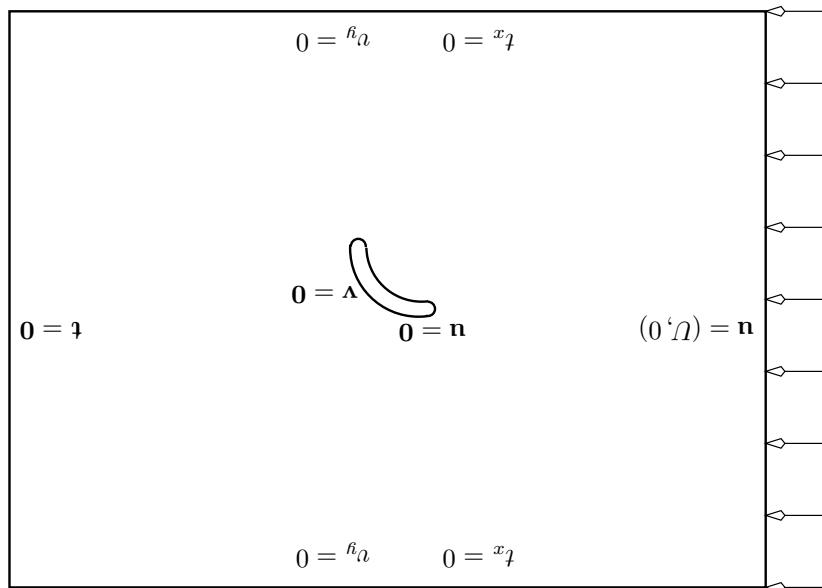


MULTI-LEVEL NEWTON method

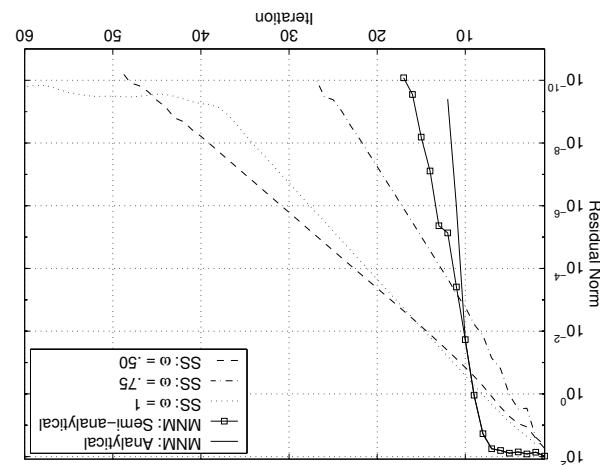
- Coupled fluids, structure, and mesh motion analyses
- Low speed flow $Re = 50$
- ≈ 4500 nodes and ≈ 8800 elements
- Iterations and CPU
 - MLN: 5 Newton \times 15 GMRES iterations \Leftarrow 161 CPU
 - Semi-analytical MLN: 8 Newton \times 15 GMRES iterations \Leftarrow 215 CPU
 - Finite difference MLN: 8 Newton \times 15 GMRES iterations \Leftarrow 2134 CPU
 - Successive substitution: 21-35 iterations \Leftarrow 350-584 CPU

Aeroelasticity

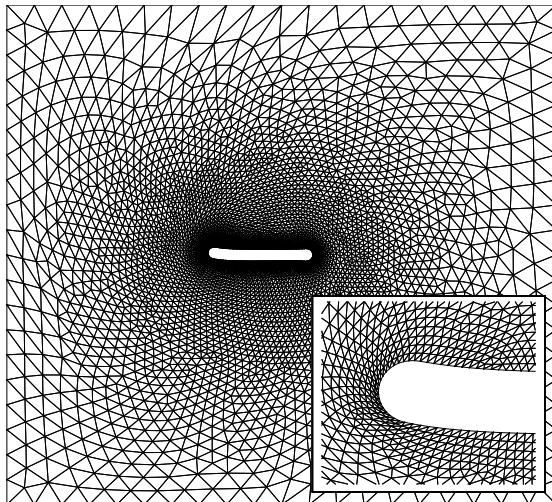
Domain and boundary conditions



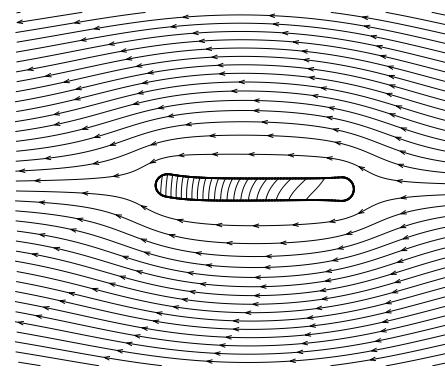
Convergence



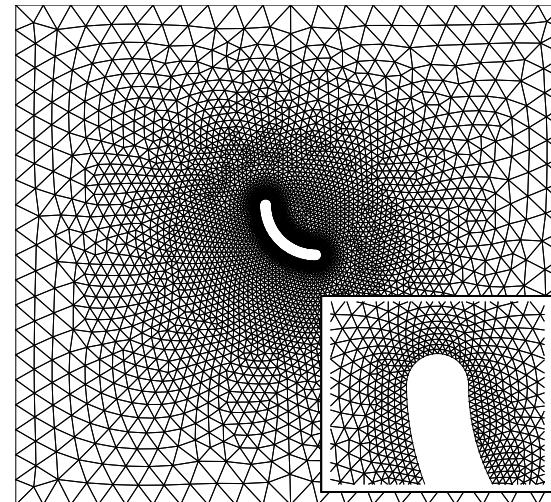
Deformed mesh

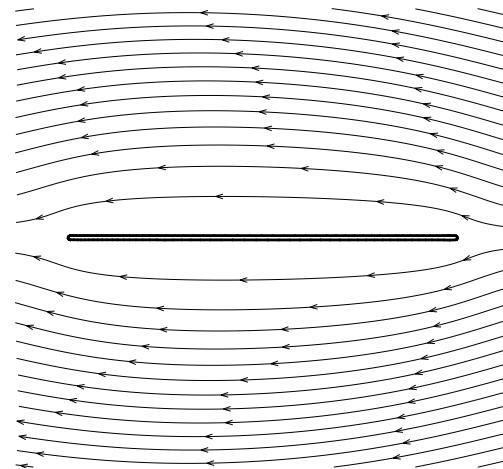
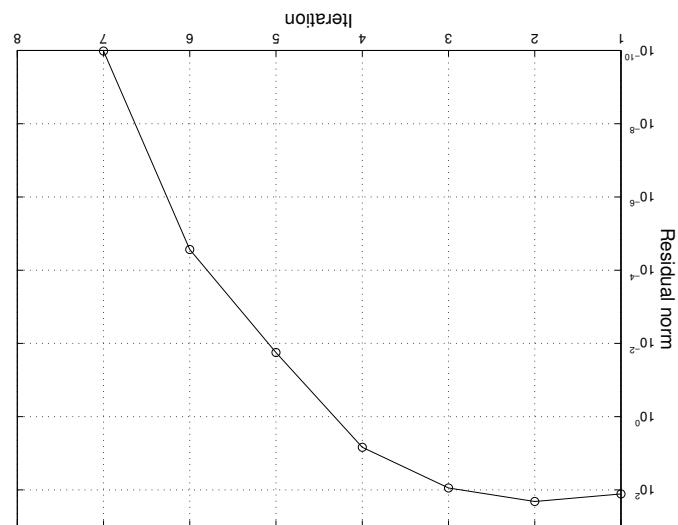


Flow field



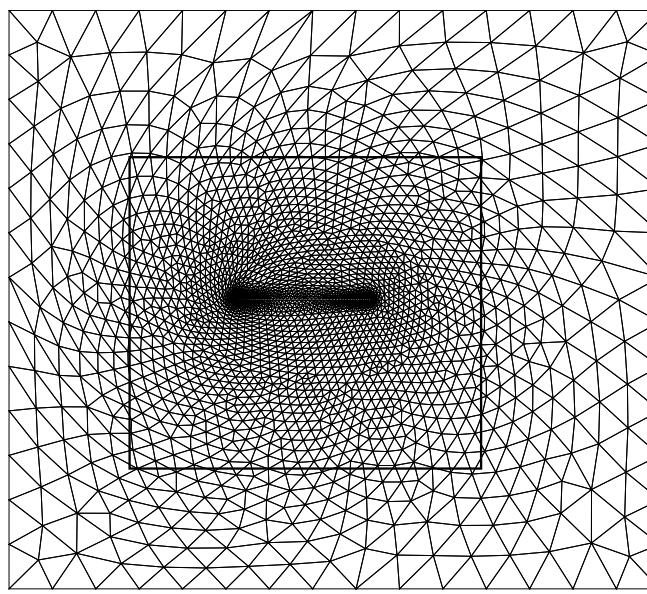
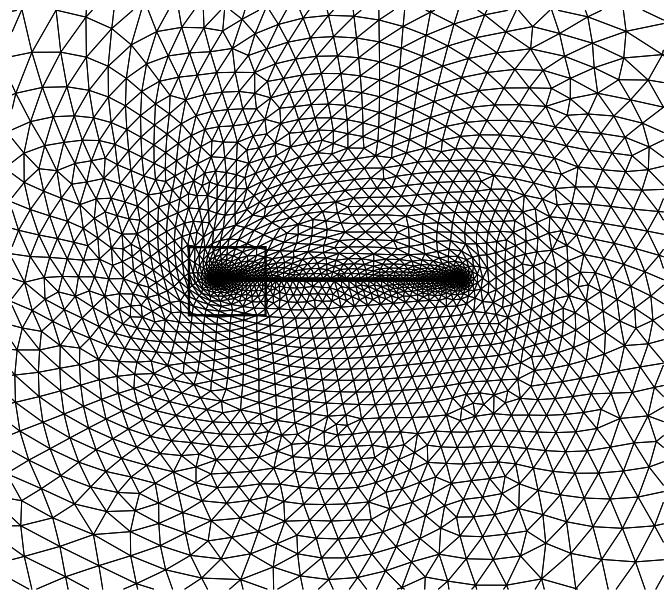
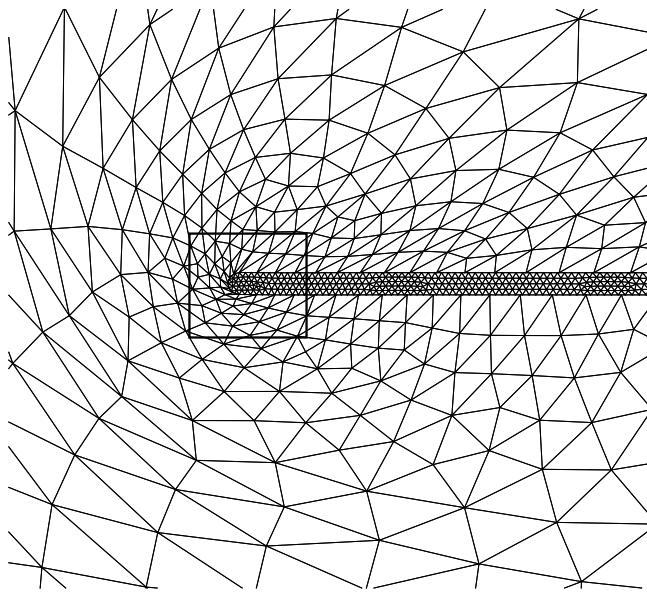
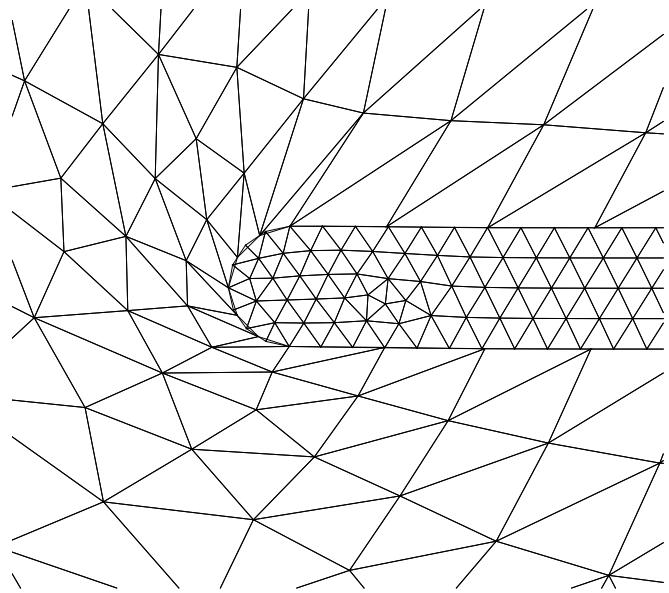
Initial mesh





- Fluid and solid meshes contain 5,662 and 2,178 elements
- Enforce interfacial momentum mass balance w/ Galerkin method
- Common refinement

Nonconforming mesh



- Local nonlinearities

- Matrix free parallelization

$$\left[\left(s \frac{\partial \mathbf{u}_I}{\partial \mathbf{R}} \right) \left[\frac{\partial \mathbf{u}_I}{\partial \mathbf{R}} \right]^{-1} \frac{\partial \mathbf{R}}{\partial \mathbf{s}} \right] \sum_u^{I=\ell} - s \frac{\partial \mathbf{u}_I}{\partial \mathbf{R}} = \\ s \left[\left(\frac{\partial \mathbf{u}_I}{\partial \mathbf{D}} \right) \sum_u^{I=\ell} + \frac{\partial \mathbf{u}_I}{\partial \mathbf{R}} \right] = \mathbf{D}\mathbf{R}\mathbf{s}$$

— Iteratively solve Newton update

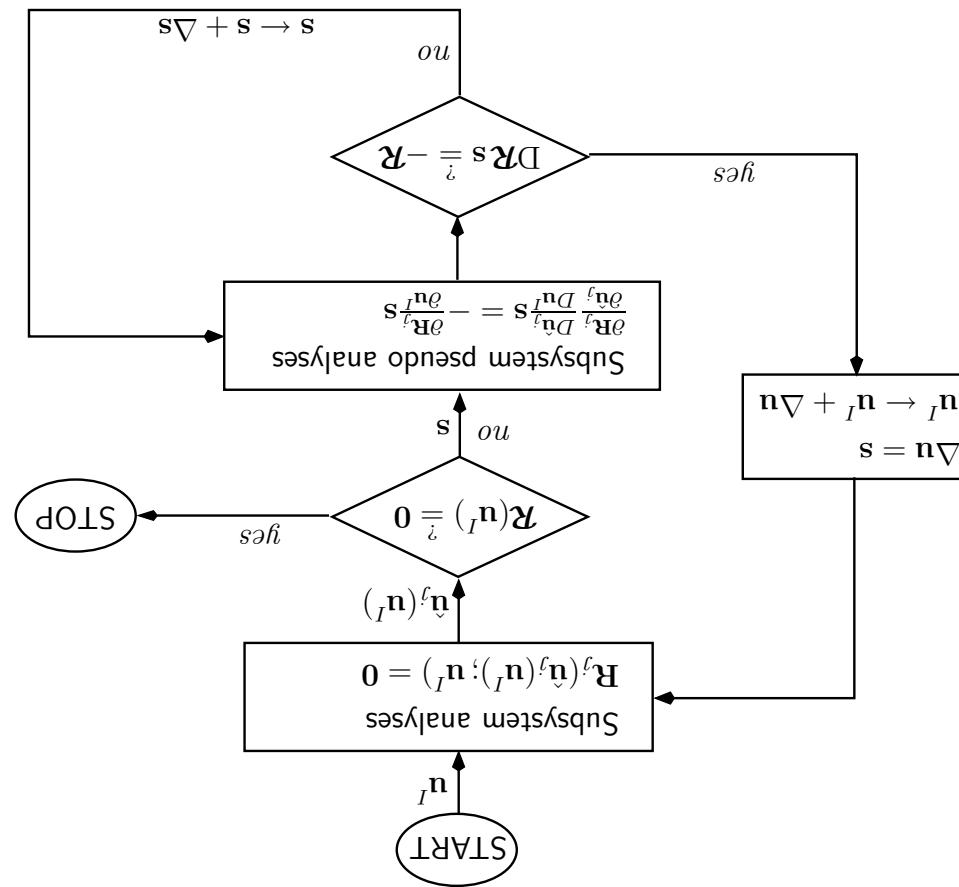
$\mathbf{0} = ((_I \mathbf{u})^u \mathbf{u}_I, \mathbf{u}_I^1, \mathbf{u}_I^2(\mathbf{u}_I), \dots, \mathbf{u}_I^n)$

- Interface analysis

— Usual balance equations:

- n subdomain analyses:

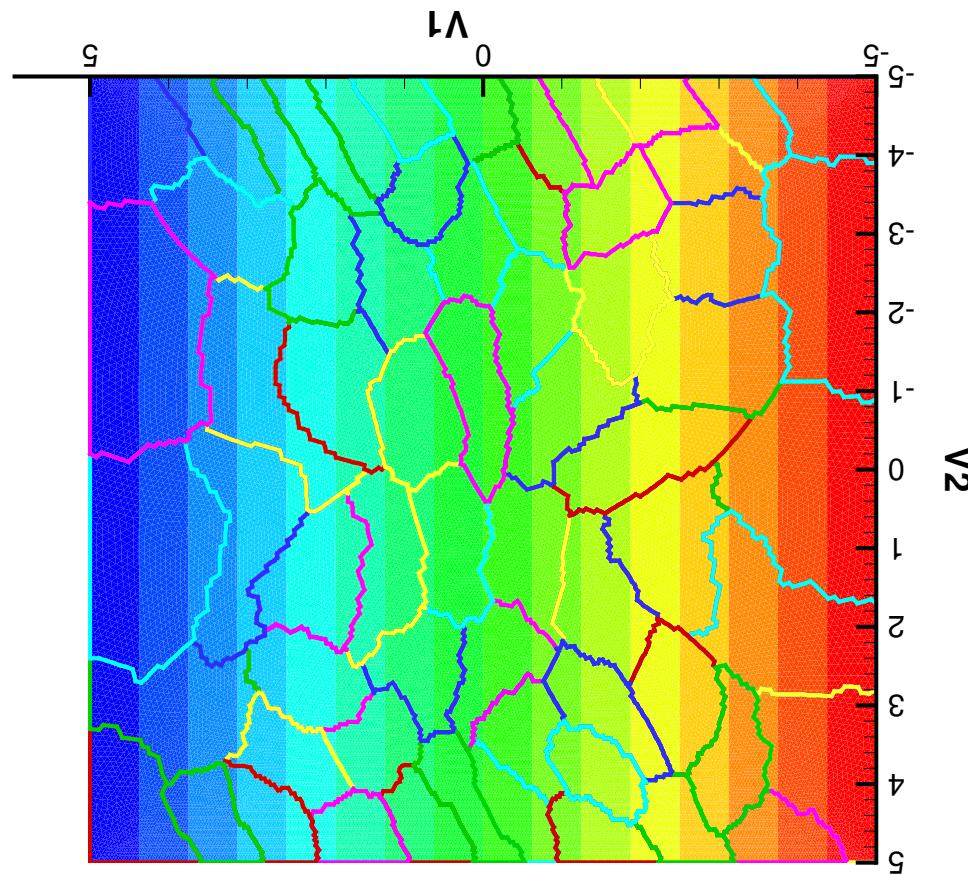
Multi-level Newton method: Domain decomposition

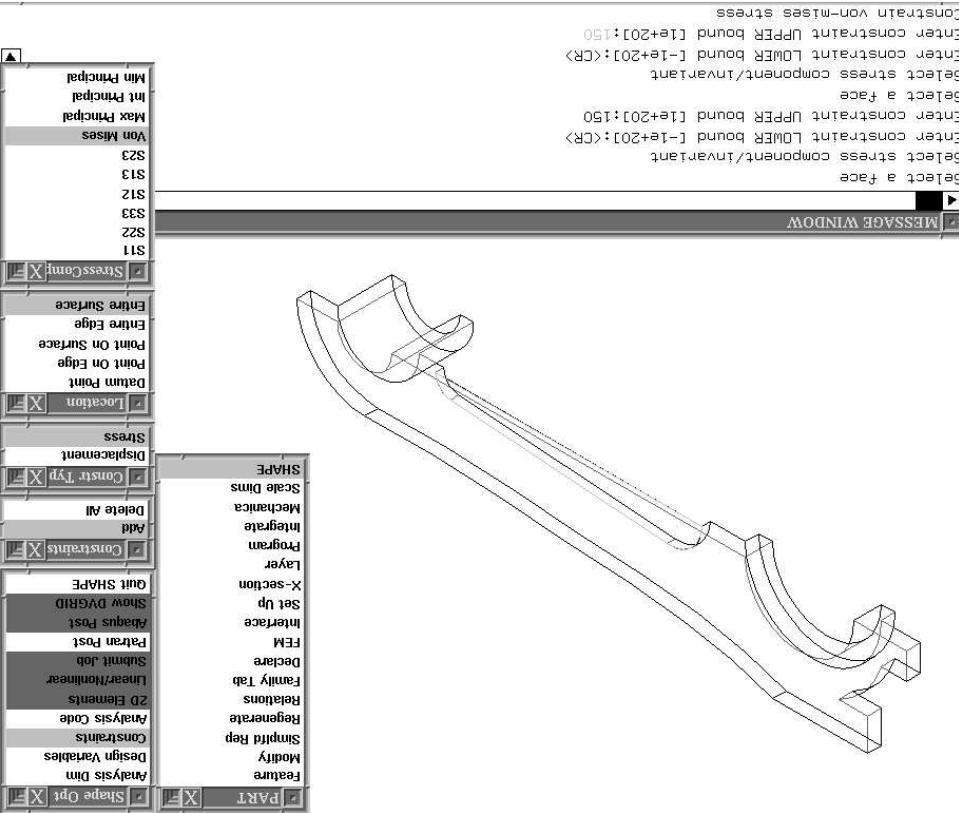


MULTI-LEVEL NEWTON method

- Linear: one outer Newton iteration
- Unstructured mesh
- $\approx 100,000$ elements $\approx \approx 60,000$ DOF
- Load balancing w/ Charm++ (PGSLib)
- Parallelization w/ MPI
- Partitioning w/ METIS (Chaco)
- Interfacial energy balance: R
- Interfacial temperature: u_I
- Steady heat conduction: B

Heat conduction example

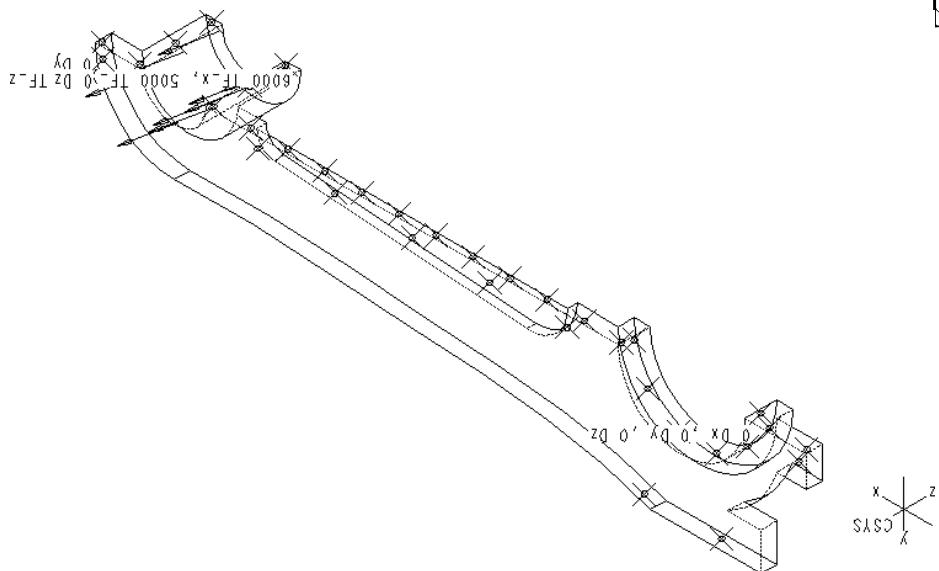




- Pro/Engineer solid modeler
- Feature-based modeling
- OCCTFE
- Pro/Develop API
- MSC/Nastran DSA
- DOT SLP

Optimization environment

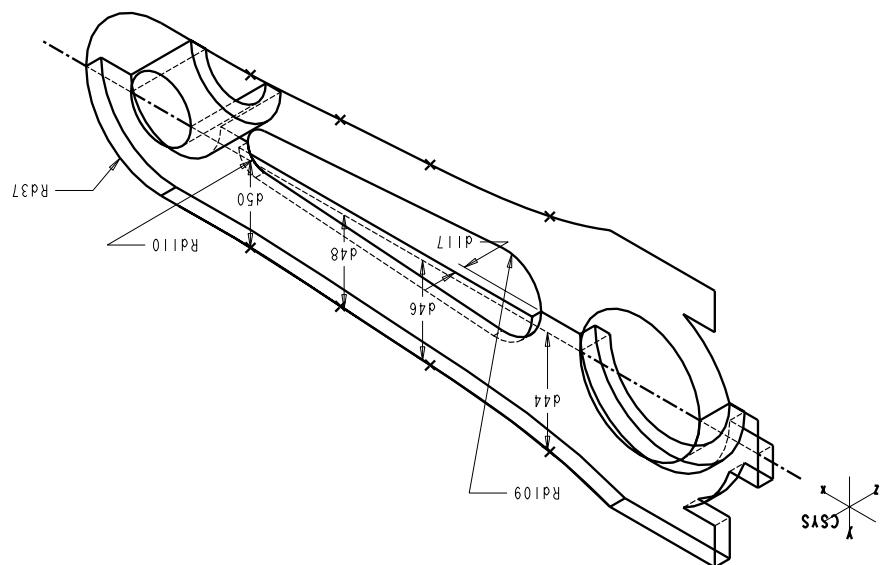
w/ S. Chen



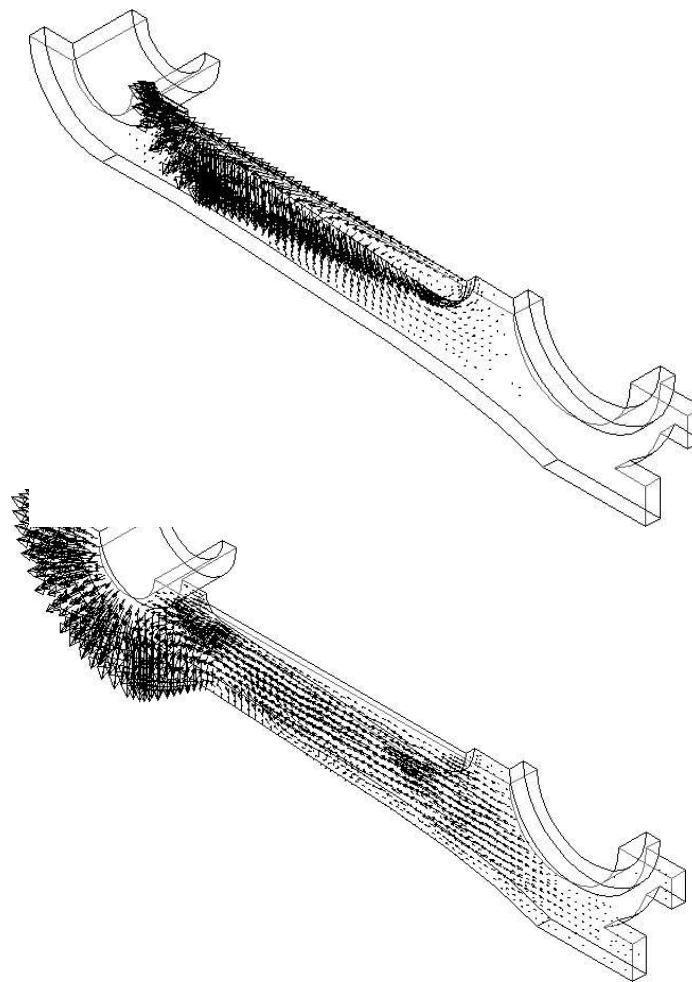
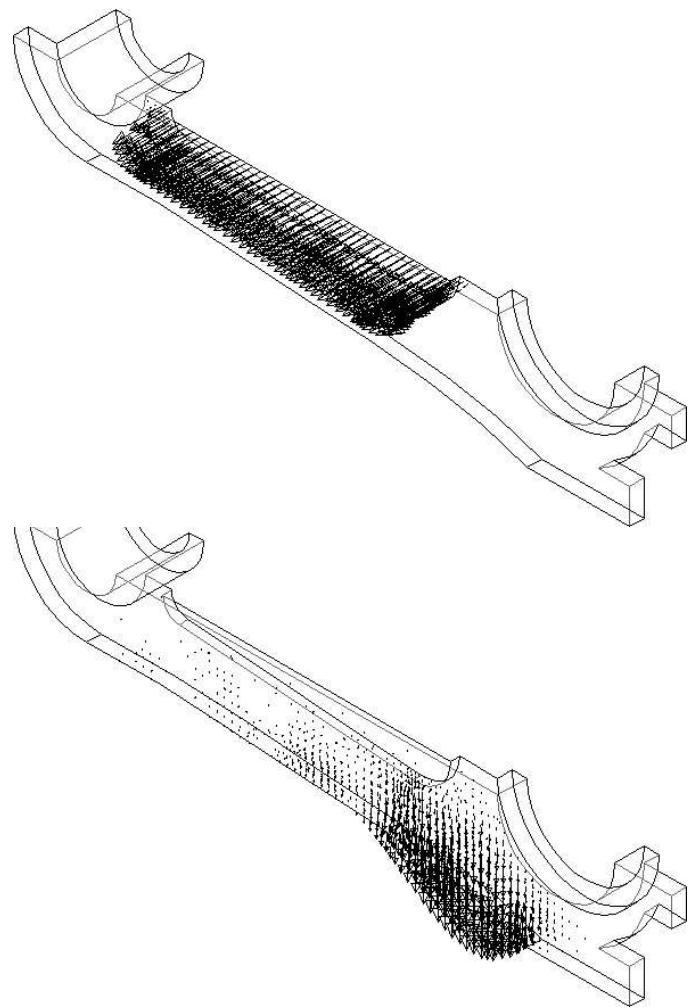
- Quarter symmetry employed
- Initial finite element mesh:
 - 2215 parabolic tetrahedrons
 - 4642 nodes
- Linear static analysis
- Steel material properties
- Boundary conditions
 - Encastre condition at crankshaft end
 - Distributed force at pin end
- Maximum von Mises stress of 79.1 MPa

Analysis model

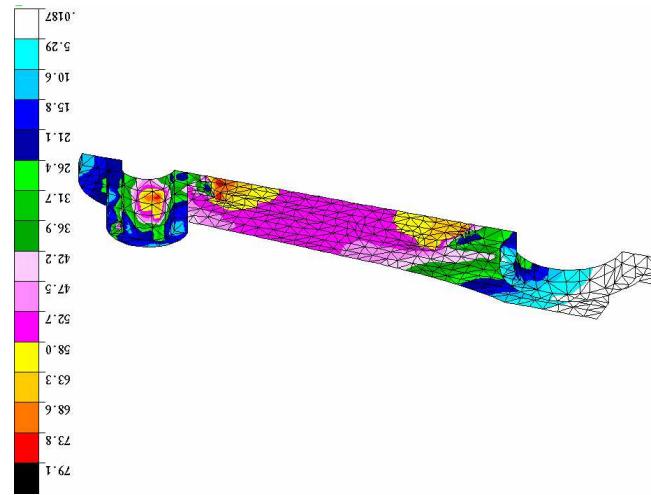
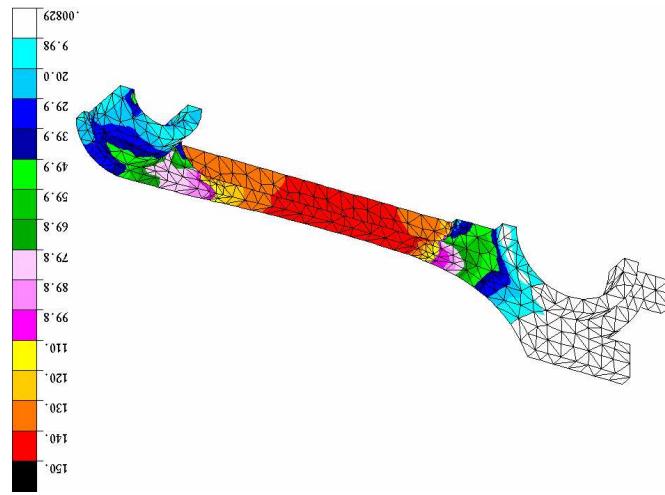
- Minimize volume/mass
- 8 Design variables
 - Neck contour
 - Web radii
 - Outer pin end radius
 - Web depth



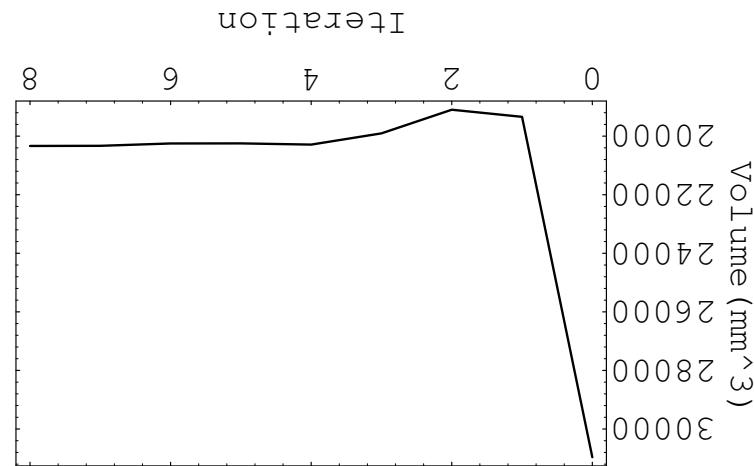
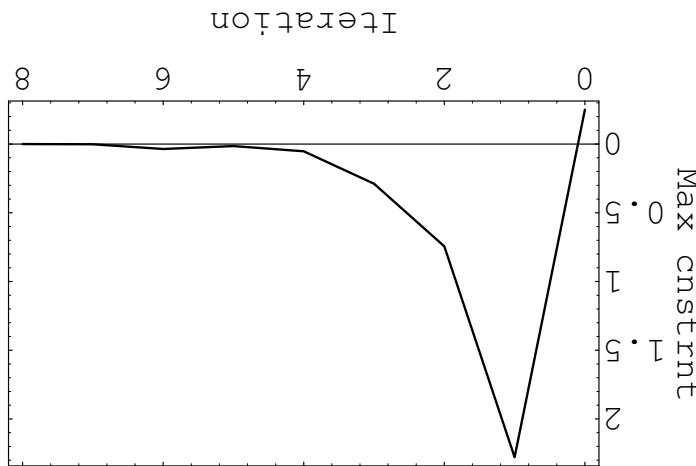
Design model



Velocity field



- Optimal design satisfies performance/geoemetric constraints
- 34.3% volume reduction
- Convergence in 8 iterations



Optimization history

DISCUSSION

- Only initial user interaction
- Update solid model dimensions each iteration

- Update finite element mesh each iteration

- Remeshing
- Laplacian smoothing

- Velocity field calculations each iteration

- Applicable to other analyses

- Mechanisms

- Crystal growth

- Controller design

- Metal forming

- Polymer processing

- Precipitation nucleation & growth

- Numerous examples

- Response surface optimization

- Design of experiments

- Reliability & uncertainty

- Many applications

[Conclusion]